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A COMPARISON OF TWO METHODS FOR PREDICTING LOSS OF LEARNING DUE TO A BREAK IN PRODUCTION

DARCOM INTERN TRAINING CENTER

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A COMPARISON OF TWO METHODS FOR PREDICTING LOSS OF LEARNING DUE TO A BREAK IN PRODUCTION

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March 1976

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#### FOREWORD

The research discussed in this report was accomplished as part of the Maintenance Effectiveness Engineering Graduate Program conducted jointly by the DARCOM Intern Training Center and Texas A&M University. As such, the ideas, concepts and results herein presented are those of the author and do not necessarily reflect approval or acceptance by the Department of the Army.

This report has been reviewed and is approved for release. For further information on this project contact Dr. Ronald C. Higgins, Intern Training Center, Red River Army Depot, Texarkana, Texas 75501.

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Department of Maintenance Effectiveness Engineering

FOR THE COMMANDER:

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Learning curves have received increased emphasis from private industry and the government in recent years. A reality that is associated with the learning curve but has received little formal attention is a break in production and the effect it has on follow-on first unit cost. Since a major activity of the government is the procurement of spare parts after initial production of a system is completed, reliable prediction techniques are needed for estimating first unit costs following a break in production.

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The ideas, concepts, and results herein presented are those of the author and do not necessarily reflect approval or acceptance by the Department of the Army.

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#### CHAPTER 1

#### INTRODUCTION

The learning phenomenon has been studied by philosophers and psychologists for centuries. In fact, Aristotle was the first to set forth laws in an attempt to explain the basis of learning. (17)\*

In Mednick's book, <u>Learning</u>, learning has been defined in terms of four characteristics. (14) These are:

- 1. Learning results in a behavioral change. This characteristic is the basic goal of any efforts at learning.
- 2. Learning is a result of practice. This eliminates behavioral changes due to illness, maturation, or motivation. Although performance may be greatly altered by these variables, learning is not.
- 3. Learning is a relatively permanent change. A task which was learned sometime in the past can be easily resumed after a little practice.
- 4. Learning is not directly observable. Performance is affected by variables other than learning. Therefore, a record of successive performance is just that, and cannot be considered an exact representation of the learning process.

As a result of studying the usefulness of learning, various methods have been developed for measuring the amount of learning acquired and projecting its effects on things to happen at some future time. The theories behind these methods of learning applications have become very useful tools in the field of industrial

<sup>\*</sup>Numbers in parentheses refer to the list of references following this report.

forecasting. One concept, which will be the foundation of this research paper, has been developed utilizing the relations between the number of units produced and the labor hours required to produce those units.

The statistical measure of this relationship between the number of units produced and their respective labor hours has become known as a learning curve. This curve also may be referred to as a progress, improvement or experience curve because variables, other than learning, contribute to determining the slope of the curve. (16) These variables are: (12)

- 1. Improved Production Methods
- 2. Direct Labor Learning
- 3. Management Learning
- 4. More Effective Procurement
- 5. Eliminating Engineering Problems
- 6. Simplification of Design

The first publication leading to the industrial application of the learning curve has been credited to T. P. Wright. His article was published in the February, 1936 issue of the <u>Journal of Aeronautical Science</u>. (11) He showed that as the number of aircraft produced increases, the cumulative average per unit cost to produce an aircraft decreases at a constant rate. This has since become known as the cumulative average theory of the learning curve. (12)

Since this first publication, the learning curve theory has been extended into many areas ranging from the setting of contract prices to production planning and control. (1) In situations where the learning curve principles can be applied, the government is also using it in evaluating contract proposals.

In the book, <u>Purchasing and Materials Management</u>, Lee and Dobler state that: (13)

The basic point revealed by the learning curve is that a specific and constant percentage reduction in the average direct labor hours required per unit results each time the number of units produced is doubled.

This means that the direct labor hours required to produce a second unit will be a certain percentage less than the labor required for the first unit; the direct labor required for the fourth unit will be the same percent less than that of the second unit; the direct labor for the eighth unit will be the same percent less than the fourth unit; and this constant percentage of reduction will continue as long as uninterrupted production of the same item continues. (12)

In using the learning curve, there are at least two pieces of data required. First, the hours required to produce the first unit must be known or determinable. This data point is arrived at by the use of production standards or historical data which has been made

available from previous builds. The second piece of data, the slope of the learning curve to be used, is arrived at by fitting a curve to the historical data, if available, or using a standard curve accepted by a particular industry.

If theoretical data is used, the accuracy of the calculations would be only as good as the assumptions used in generating them.

For this reason, the best results are obtained when actual or historical data is used with the learning curve.

Very often the government has access to historical or actual data which its contractors have accumulated during past contracts of like or similar items. This is especially true where follow-on contracts for spares or replacement parts are concerned. A phenomenon associated with these types of contracts which has received very little formal attention is the effect a break in production might have on a follow-on contract's first unit direct labor hours. More and more concern is being directed to this area of learning curve application because the problem of pricing the first item to be produced following a break in production exists in numerous contract negotiations.

Two methods, one by Allen A. Pichon and Charles L. Richardson of the USAF and another by Robert Blair Ilderton of the Defense Contracts

Audit Agency, have been developed for dealing with follow-on contracts which have experienced production breaks. The objective of this research paper will be to examine these two methods to see how they compare in predicting the first unit cost following a break in production.

#### CHAPTER 2

# LEARNING CURVE BACKGROUND AND RESEARCH JUSTIFICATION

In the previous chapter mention was made of the learning curve as a tool for estimating labor hours for production contracts. This chapter will provide an introduction to the learning curve for the benefit of those who are unfamiliar with its principles and a more detailed justification for this research paper.

## The Learning Curve

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The kind of curve that expresses the learning phenomenon is called an inverse variation. The formula used to express this relationship is  $Y=KX^C$ , where

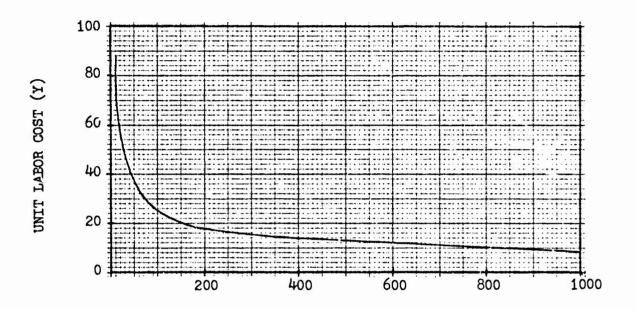
- Y = the number of direct labor man hours required to produce the Xth unit.
- K = the number of direct labor man hours required to produce the first unit.
- X = the unit number
- c = log B/log 2 where B equals the learning curve factor (.90, .85, .77, etc.)

Also, since the inverse variation, in general, means that the dependent variable (Y) gets smaller as the independent variable (X) gets larger, this relationship is also referred to as an exponential (log-linear) equation. For a given learning curve, K and c are constants where K can assume any positive value and c is a negative constant between 0 and -1. (15)

It is important to understand K and c since they control the vertical position and rate of decrease of Y. Due to the fact that  $1^{n} = 1$ , it is evident that Y = K for X = 1. Thus, the magnitude of the vertical height of Y, the dependent variable, is determined by K. For increases in X the rate of decrease in Y is controlled by the size of c which means that as c approaches 0, Y approaches a horizontal line K units high and tends to decrease very little for increases in X. Yet, the rate of decrease of Y grows larger as c approaches -1. (15)

As can be seen from Figure 2-1, the greatest absolute decreases in Y values occur at the lowest range of X values for the curve described by Y = KX<sup>C</sup>. Upon closer observation of the curve, it becomes apparent that, for a unit change of X, the absolute decrease in Y gets smaller as X increases. In fact, each time the value of X is doubled, the value of Y will decrease by a constant proportion. The amount of decrease will depend upon the value of the constant c. For example, on a 90 percent curve the cost of the 100th item of a production run will be 90 percent of the cost of the 50th item, the cost of the 50th item will be 90 percent of the cost of the 25th item, etc.

By taking the logarithm of the learning curve equation and remembering that log(AB) = log(A + log(B)) and  $log(Z^y) = y log(Z)$ , then



CUMULATIVE UNIT NUMBER (X)

Figure 2-1. 90 Percent Learning Curve On Arithmetic

Grid For Y = KX<sup>C</sup> Where c = Log .9/Log 2 (15)

 $Y = KX^C$  becomes

$$\log Y = \log K + c \log X \qquad (Eq. 2-1)$$

If Equation 2-1 is compared to the standard linear equation, Y = A + BX, the similarity becomes readily apparent. Thus, for logarithmic values of Y, K, and X, Equation 2-1 can be plotted as a straight line having a negative slope.

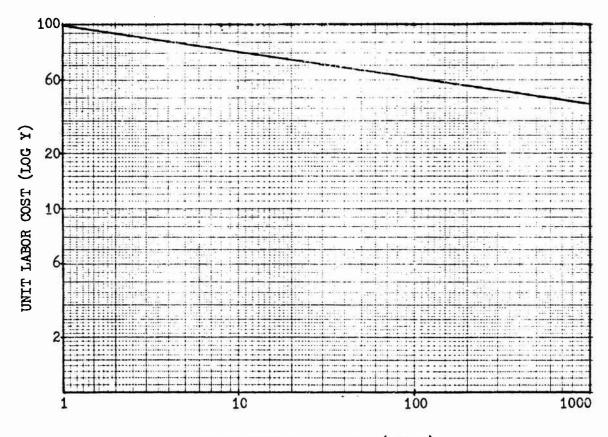
Due to the tediousness of transferring data to logarithmic values, log-log graph paper should be used to graph Equation 2-1. Figure 2-2 represents the 90 percent learning curve plotted on logarithmic grid with both the X and Y axes subdivided logarithmically. Thus, an arithmetic value plotted on log-log paper corresponds to logarithmic values plotted on arithmetic graph paper. (15)

The slope of the learning curve is another interesting characteristic to observe. The mathematical definition of slope is given as a logarithmic function of c. Since Y is customarily expressed as a function of X where  $X_N$  differs from  $X_{N-1}$  by a factor of 2, the equation  $Y = KX^C$  can be used to express the slope as follows:

Slope = 
$$\frac{Y_{2X} = K (2X)^{c}}{Y_{X} = KX^{c}}$$

or

$$Slope = S = 2^{C}$$



CUMULATIVE UNIT NUMBER (LOG X)

Figure 2-2. 90 Percent Learning Curve On Logarithmic Grid For

Log Y = Log K + c Log X, Where c = Log .9/Log 2 (15)

Thus, c can be expressed as a function of S by taking the log of both sides of the above equation. (15)

$$log (slope) = log S = c log 2$$
 (Eq. 2-2)

Therefore.

$$c = \frac{\log S}{\log 2}$$
 (Eq. 2-3)

For the 90 percent learning curve, c can be expressed as

$$c = log .9/log 2$$

Therefore, by knowing the type of learning curve involved,  $Y = KX^{C}$  can be used as a model for predicting various values of Y.

### Justification for Research

One of the assumptions that is important in the application of the learning curve is that production runs will be stable and will not encounter a break in production. In real world situations however, production runs may not parallel the above th oretical assumption.

In fact, a problem currently facing contract negotiators who use the learning curve as an evaluating tool is that of relating the effects of a break in production to the learning curve. In his unpublished study, "Production Break and Related Learning Loss,"

George Anderlohr, a former employee of Defense Contracts Administration Services (DCAS), discussed an example of a break in production. (3)

The following quote from his study is pertinent to the problems faced by government negotiators:

The production break is the time lapse between the completion of a contractual requirement for the manufacturing of certain units of equipment and the commencement of a follow-on order for identical units of equipment. This time lapse disrupts the continuous flow of products. This could, in smaller shops, include a condition where the follow-on order was received prior to the delivery of the last units of the first order. An example of this would be the completion of circuit board assemblies, and all personnel had been moved into the final assembly area. Thus, the circuit board assembly line would have to be reestablished to accommodate the new order. (3)

In establishing a new assembly line, lost learning would be implied.

It is the above learning loss that negotiators are concerned with when negotiating a follow-on contract which has experienced a break in production. Mr. Anderlohr further emphasized this problem in the above mentioned study as follows:

A major problem with the application of the improvement (learning) curve has always been that it addresses itself to a perfect environment which rarely exists. A major condition for this perfect environment is an uninterrupted production cycle (one lot of identical units following another). When plotting actual labor hours on a curve, it has been long noted that any interruption in the orderly and continuous flow of work from one work station to another is accompanied by an increase of labor hours when production is resumed. This has been commonly referred to as start up costs which relates directly to loss of improvement.

In the real world of government procurement there is, almost always, a break in the production cycle. There has been no established reliable method of compensating for the loss of improvement resulting from a break in production. General Electric Cost Accounting Service Bulletin No. PC-5 recommends a fifty percent loss of learning for a three to six month break and a seventy-five percent loss for a twelve month break. This is such a general approach that

it would be extremely difficult to support in cost negotiations. Because of the lack of guidance, most cost analysts take almost arbitrary positions ranging from the use of unexpected percentages, as mentioned above, to the position that no learning was retained after a production break. The total loss of learning is usually based on a common misconception that learning or improvement is directly related to personnel know-how only.

Negotiators and Cost Analysts facing their counterparts across a negotiation table are frequently plagued with the recurring problem of estimating loss of improvement (learning). (3)

The problem facing contract negotiators is the determination of how much learning has been lost and what point should be used upon reentering the learning curve to determine unit cost. This problem of determining the effects of a break in production on a follow-on contract has not been researched extensively as of yet. In performing the literature search on this topic, two models were found which have been developed for the purpose of predicting lost learning. The purpose of this research paper will be to make a comparison of the results of these two methods when predicting the cost of the first unit produced following a break in production.

#### CHAPTER 3

## DESCRIPTION OF MODELS

This chapter presents the results of a literary search that this author has made on the subject of determining effects of lost learning due to a break in production. An attempt will be made to briefly describe two mathematical models, both designed for use on the computer, that have been developed by two separate sources.

#### Model 1:

The United States Air Force (USAF), which relies a great deal on the learning curve as an estimating tool in many of its contract negotiations, prior to 1974 had no formal method for handling loss of learning due to a break in production. Thus, the USAF supported a study by Allen A. Pichon, Jr. and Charles L. Richardson which resulted in a mathematical model for predicting the first unit costs following a break in production by use of step-wise regression techniques. (15)

In developing their method, Pichon and Richardson used the Newtonian approximation method, simple calculus and the computer to transform collected data into meaningful data, provided the number of units in a lot, the cumulative direct labor hours involved, and the break in production are known. Thus, a means of approximating the first

unit direct labor hours, the last unit direct labor hours and the learning curve for a lot was formulated.

The above authors formulated a mathematical model that was developed by using step-wise regression techniques. These techniques were used based on the objectives of regression analysis listed:

- 1. The first purpose of regression analysis is to provide estimates of values of the dependent variable from values of the independent variable(s)...
- 2. A second goal of regression analysis is to obtain a measure of the error involved in using the regression line as a basis for estimating. For this purpose the "standard error of estimate" or its square, the "error variance around the regression line" are calculated...
- 3. The third objective, which...(is)...classified as correlation analysis, is to obtain a measure of the degree of association or correlation that exists between... two variables. The coefficient of determination, calculated for this purpose measures the strength of the relationship that exists between...two variables (15)...

The model resulting from the above study was of the form:

$$Y = A_0 + A_1 X_1 + A_2 X_2 + A_3 X_3$$
 (Eq. 3-1)

where

Y = the calculated independent variable (first unit cost after a break in production)

An = the regression constant

 $A_1$  = regression coefficient for  $X_1$ 

 $\mathbf{A}_2$  = regression coefficient for  $\mathbf{X}_2$ 

 $A_3$  = regression coefficient for  $X_3$ 

X<sub>1</sub> = learning curve factor

X<sub>2</sub> = last unit direct labor hours for the lot(s)

 $X_3$  = the length of a break in production

In approximating values for the variables, one must determine the learning curve for the production process and the number of hours to produce the first unit. If the recorded data reflects actual hours for the unit, then it is easy to determine both the learning curve and the first unit total hours. On the other hand, if the data recorded only reflects the cumulative hours for the units produced, and the number of units in the lot, then the learning curve and the direct labor for the first unit must be calculated by the use of some approximation technique. One such technique used by Pichon and Richardson in developing this model is described here for information to the reader.

It is generally accepted that the area under a curve "f" is represented by the integral of the function "f" bounded by an interval (a, b). This area, in terms of the learning curve, is represented by the cost of a particular lot or the total direct labor hours of a lot. Thus, the learning true can be determined by integrating  $Y = KX^C$ ; therefore,

$$D_i = {}^{t} {}^{+} {}^{.5} KX^{c} dx$$
 (Eq. 3-2)

where D; equals the direct labor hours of a particular lot of size "t".

This integration then is actually an approximation of a step function since direct labor hours is a discrete variable. After expanding

$$D_i = \frac{K}{c+1} \left[ (t + .5)^{c+1} - (.5)^{c+1} \right]$$
 (Eq. 3-3)

for a lot of size L, D can be expressed as:

$$D_{L} = \frac{K}{c+1} \left[ (L + .5)^{c+1} - (.5)^{c+1} \right]$$
 (Eq. 3-4)

and the direct labor hours for both lots, ie. the sum of lot 1 and lot 2 can be expressed as:

$$D_{M} = \frac{K}{c+1} \left[ (M + .5)^{c+1} - (.5)^{c+1} \right]$$
 (Eq. 3-5)

Expressing Equations 3-4 and 3-5 in another form:

$$\frac{c+1}{K} = \frac{1}{D_L} \quad \left[ (L + .5)^{c+1} - (.5)^{c+1} \right] \quad (Eq. 3-6)$$

and

$$\frac{c+1}{K} = \frac{1}{D_{M}} \quad \left[ (M + .5)^{c+1} - (.5)^{c+1} \right] \quad (Eq. 3-7)$$

Solving Equations 3-6 and 3-7 simultaneously:

$$D_{L} \left[ (M + .5)^{c+1} - (.5)^{c+1} \right] - D_{M} \left[ (L + .5)^{c+1} - (.5)^{c+1} \right] = 0$$
(Eq. 3-8)

For Equation 3-8, the following information is determinable from actual data collected: (15)

 $D_{t}$  = Direct labor hours for the first lot

 $\mathbf{D}_{\mathbf{M}}$  = Direct labor hours for two or more consecutive lots

L = Number of units produced for the first lot

M = Number of units produced for two or more consecutive lots

We see that the only variable that is not defined is c and Equation 3-8 can be used to solve for this unknown. Since c equals log B/log 2, the learning curve factor, B, can also be determined. However, calculating the value for c is a long and tedious process whereby a value for c must be selected and plugged into Equation 3-8 to see if it equals zero. Chances are very slim that this first selection will be the exact value that satisfies the equation. Therefore, other values (one higher and one lower than the first) must be selected in efforts to find an interval in which c must lie. Once this interval is determined, the selection process must be continued until a value for c is found for which Equation 3-8 is satisfied. Since such a process of elimination could be quite inefficient because of the selections made and the accuracy desired, other methods for approximating the learning curve can be used.

One such method is to make use of learning curve table; which have been generated to assist in applying the learning curve as an estimating tool. These tables contain factors for various learning curves which can be used to establish the amount of time required to produce any unit in

a production lot. In order to use such tables, one needs to know or be able to determine the lot size, the cumulative or unit costs from previous productions and the learning curve for the items produced. A portion of such tables for an 8% learning curve is shown in Figures 3-1A and 3-1B. (9)

To illustrate how learning curve tables can be used, suppose that you have built a lot of 25 units and have gotten a follow-on order for two option quantities of 25 units each and you need to estimate the cost of these options. You have recorded actual hours for the 25 units at 8500 hours and an 85% curve was experienced.

From the example,

Hours for Unit 1 = 
$$\frac{\text{Total Cumulative Hours}}{\text{Cumulative Factor from}} = \frac{8500}{14.800727}$$
 (Ref. Fig. 3-1A)  
Learning Curve Table  
for N = 25

= <u>574.2961 Hours</u>

one can determine the total cost for the two follow-on options.

#### Option 1:

Total hours for 25 additional units = Total hours for 50 units— Total hours for 25 units

Total hours for 50 units = Hours for Unit 1 x Cumulative Factor for 50 units

= 574.2961 x 25.51311 (Ref. Fig. 3-1A)

= 14,652 Hours

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0.98	C.A.	.00000000 .93000000 .88245886 .84674414	.79477794 .77478223 .75744145 .74216594 .72854041	.71626166 .70510184 .69488547 .68547471	.66864988 .66107281 .65396695 .64726087	.62933581 .62933581 .62395056 .61881994	092397 047547 004526 963201 923455 885178	.58126570 .57782442 .57449638 .57127492 .56315393 .56512780	.55656891 .5538440 .55125254 .54621300 .5412495 .53111821 .5366681
	CUM TOTAL	1.00000000 1.86000000 2.64737659 3.38697659 4.09152383	4.76864770 5.42347559 6.05953159 6.67949349	7.87887835 8.46122208 9.05351117 9.5964596 10.15138996	10.69839812 11.23823787 11.77146510 12.2983361 12.81941974	13.33500015 13.84538800 14.35086300 14.85167861 15.34806541	5.8402340 6.3283734 6.8126334 7.2032835 7.4703635 8.2440535	19.18176824 19.64663642 20.10737351 20.56585733 21.02165076 21.9254538 22.37359530	22.81932548 23.26272463 23.70335937 24.14.279291 24.57758536 25.014.2°386 25.4697287 25.3064767 26.3064763
•	INO	.0000000 .8499999 .77291483 .7225000	.65697761 .63365646 .61412500 .59739734 .58782040	.55994060 .55843096 .54404849 .53860799	.52200624 .51463874 .50778774 .50139120	.48976249 .48444951 .47942661 .47466632 .47014480	658412 617372 578167 540654 504705 470205	.44051555 .43447990 .43447990 .42845569 .42845569 .42108773	.41865687 -41629811 -41400769 -41178209 -40361806 -40751262 -40546292 -401346637 -401346637 -401362308
85.0		.00000000 .9250000 .87430494 .83635370	.78134391 .76024571 .74198062 .72591581 .71160627	.69872757 .68703619 .67634483 .66656648	.64894140 .64106124 .63363827 .62661783	.61390666 .60802224 .60243113 .59710760	5871757 5825298 5780758 5737995 5696885 5657315	.55823939 .554¢8655 .55125207 .5471076 .54471076 .53856603	.53277609 .53000281 .52730525 .52467972 .52212279 .51720215 .51720215 .51720215
	CUM TOTAL	1.00000000 1.8499999 2.62291484 3.34541483 4.03108589	4.68206350 5.32171997 5.93584497 4.53324232 7.11606272	7.68600333 8.24443430 E.19248279 9.33109379	10.38306238 10.89?70113 11.40548887 11.90680007	12.89203990 13.37648942 13.85591603 14.83058238	5.7265683 6.7283056 6.1861224 6.6401879 7.090658 7.5376790	18.42189989 18.85934283 19.29382274 19.72544231 20.15429831 20.56544231 20.56544231 20.56544231 21.6564309	21.84381996 22.26011807 22.674;2576 23.08590785 23.49552593 23.4952593 23.4952593 23.4952593 24.30850147 25.11134;337 25.5131;145
	z		97860	123211	16 17 19 20	22 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5		4 W W W W W W W W W W W W W W W W W W W	11444 44440 000000 00000

FIGURE 3-1A 85% LEARNING CURVE TABLE

							•	
AGE 502	UNIT	.45386581 .45209858 .45037169 .44868320	.44541705 .44288777 .44228777 .44228777 .43928747 .43928747 .435000000000000000000000000000000000000	.42965361 42965353 42837656 42712143 42588896	42467695 -42348529 -4231332 -42116548 -42002621	.41390994 .41781119 .41566420 .41461505 .41356153 .41256321 .41256321 .41057661	.40022973 .4072811 .40582874 .40402079 .403134649 .40139649	35570 3550 3550 3550 3550 3550 3550 3550
97.0 P	C.A.	.55800983 .55597308 .55398060 .55203065	.54825187 .54464001 .5476446 .5413646 .541368 .5334444 .5345499 .5345499	.53142989 .52941084 .52841769 .52594964	.52408578 .52208855 .52131356 .51995014	.51731560 .51602334 .513475034 .51226007 .51226007 .500844086 .50085478	.50520129 .50408042 .50408044 .50180283 .50208550 .497654183 .49765495 .47663093	4946203 4936332 4926579 4916940 4907415
•	CUM TOTAL	28.45850140 28.91059998 29.36097168 29.807655:8 30.25668763	30.70210467 31.14594C60 31.02894972 32.02894972 32.46828519 32.46828519 32.46828519 32.46828519 34.2114185 34.2114185	35.07437302 35.50402655 35.93240312 36.3595205 36.78541401	37.21009097 37.63357625 38.05588958 38.47705037	39.31598622 39.73379742 40.56617105 40.9666105 40.98080610 41.39438763 42.21851056 42.62908117 43.03867674	43.44731093 43.85499708 44.26174819 44.66757694 45.47651647 45.47765109 46.28191104	7.483537 7.8824246 0.2604744 8.6777131
	UNIT	.42505717 .42326501 .42151432 .41980340	.41649448 .41489353 .41332642 .411332642 .41028866 .40081565 .40737176 .4059594	.40106744 .40055463 .39926547 .39799918	.39553237 .39433049 .39314375 .39198656	.38971854 .38661161 .35644736 .39539308 .39833276 .3833278 .3833278 .3833236 .3833236	.37937587 .37642273 .37652887 .37655887 .3764149 .37787092 .37297092	370403 369568 366745 367931 367931
86.0	C.A.	.53251760 .53041659 .52835183 .52635148	.52245723 .52057015 .51872113 .51690876 .51513176 .515188887 .51000078 .50835337	.50514679 .50358572 .50205159 .50054359	.49760275 .49616841 .49475718 .49336839	.49065556 .4893331 .48673436 .48673436 .4867247 .4867247 .4867247 .4817807 .4817807 .4817807 .47140546	47824232 47709497 47594301 47374380 4775383 47753186 47753186 47753186	464105 4664105 4664105 4664293 4634563
ů,	CUM TOTAL	27.15839750 27.58166251 28.00317684 28.42298024 28.84111087	29.25760536 29.67249490 30.08582532 30.49761719 30.90790585 31.72609327 32.1300492 32.3361646 32.93762115	33.33968859 33.74024323 34.13950870 34.53750789 34.93426295	35.32979533 35.72415882 36.11727457 36.50926114 36.90010449	37.28982303 37.67843465 28.06595670 38.45249606 38.83779915 39.22215191 39.22215191 39.36974787 39.98774913 40.36912140	41.12883987 41.80726261 41.88474546 42.26130134 42.63594234 43.30553157 43.75850248 44.1300644	4.8122517 5.2418267 5.418267 5.4185035 6.3456309
	GNIT	.39777192 .39596503 .29420054 .39247668 .39079177	.38914428 .38753270 .38595565 .38549182 .38289795 .37856747 .37856767 .3785469	.37443821 .37312032 .37182649 .37055594 .36930791	.36808171 .36587664 .36569706 .36452734 .36452734	.36225514 .36114655 .36104559 .35898176 .35792457 .35585834 .35484841 .35485834	.35190653 .3509534 .35091677 .34817535 .34817535 .3451677 .3451677 .34564348	3429464 3421141 3412924 3404810 3396796
85.0	C.A.	.50805654 .50590093 .50379337 .50173195	-49580692 -49280692 -49205698 -4923169 -48845318 -48467060 -4848818 -48330222 -48164805	.48002366 .47842809 .47686041 .47531977	.47231625 .47085181 .46941127 .46799392 .46653909	.4522614 .4533/440 .456254345 .451231254 .4594119 .45617934 .45617934 .45617934 .45617934	.45257580 .45140773 .45025554 .44799723 .44687039 .44511958	426037 415655 405403 395276 385271 375386
	CUM TOTAL	25.91088337 26.30684841 26.70104896 27.09352564 27.48431742	27.87346170 28.26099440 28.64695006 29.C3136188 29.C1426183 30.17564819 30.55419269 30.93134240	31-68156161 32-05468194 32-42650843 32-79706438	33.53445401 33.90133065 34.26702271 34.63155006 34.79443195	35.35718710 35.7183335 36.07838924 36.43737100 36.43737100 37.15217917 37.852803751 37.852803751 37.85239353	39.27247263 39.27247263 39.62248747 39.97157616 40.31975155 41.35392655 41.70356664	2.3902998 2.3902998 2.7324146 3.0737055 3.4141875 3.7538671
	z	55.922	8688 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6 6 8 4 6 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	12222	977776 977776 977776 9893 9893 9893 9893	300000 H.264	100984

PIGURE 3-1B

Thus,

Total hours for 25 additional units = 14,652 - 8500 = 6,152 Hours

or

$$\frac{6152 \text{ Hours}}{25 \text{ additional units}} = \frac{246.08 \text{ Hours/Unit}}{25 \text{ Additional units}}$$

#### Option 2:

Total hours for second 25 additional units = Total hours for 75 units Total hours for 50 units

Total hours for 75 units = Hours for Unit 1 x Cumulative factor for 75 units

= 574.2961 x 34.9949 (Ref. Fig. 3-1B)

= <u>20,097 Hours</u>

Thus,

Total hours for second 25 additional units = 20,097 - 14,652 (from Option 1)

= 5,445 Hours

or

Since the learning curve is an exponential function, Pichon and Richardson decided to look at transformations of the linear model in efforts to improve its predictive capability. Their transformed model came as a result of developing twenty-five different models, one of which was chosen to replace the original model. The following model was chosen: (15)

$$Y = e^{A_0 + A_1 X_1 + A_2 (\ln X_2)}$$
 (Eq. 3-9)

or

$$\ln Y = A_0 + A_1 X_1 + A_2 (\ln X_2)$$
 (Eq. 3-10)

where

Y = the calculated dependent variable (first unit cost after a break in production)

A<sub>0</sub> = the regression constant

 $A_1$  = regression coefficient for  $X_1$ 

 $A_2$  = regression coefficient for  $X_2$ 

X<sub>1</sub> = last unit direct labor hours for the lot(s) preceding the
 break in production

X<sub>2</sub> = learning curve factor

This transformed model is the one selected by Pichon and Richardson in making their analysis because it yielded a standard error of estimate of only 1.5696 at a .05 level of significance compared to a standard error of 25.6335 for the basic model. This model was also selected because a break of as much as 23 months was shown to be statistically insignificant in estimating the cost of a production lot following a break in production. For these reasons, the natural logarithm version (Eq. 3-10) of this transformed model will be used in making the comparisons in Chapter 4.

In order to determine values for the unknowns in the above equations, stepwise regression and the learning curve tables can be used. Pichon and Richardson calculated values for the regression constant and coefficients  $(A_0, A_1, A_2)$  in reference 15. They were calculated by applying stepwise regression to data that came from small cost items, items which took less than ten hours to produce. Those constants produce reasonably good results for items which require less than 50 direct labor hours (DLH). However, as the DLH for the last unit of a previous lot,  $X_1$ , get larger, the estimate for Y increases to the point that the estimate for the next unit to be produced takes more time than was required to produce the first unit of lot one. At this point new constants should be calculated so that the equation will better represent direct labor data for the item being evaluated.

If the unit DLH are available, then the above constants can be calculated by applying stepwise regression using actual data. The reader is referred to Wiley's <u>Applied Regression Analysis</u> by Draper and Smith for a procedure in applying stepwise regression. If, however, only the total DLH are available, then the DLH for each unit must be estimated to provide data for calculating the regression constants and coefficients.

Since the learning curve can be determined as stated previously, the unit DLH for each unit produced can be estimated as follows.

First Unit DLH = 
$$\frac{\text{Total DLH}}{\text{Cum. F.}}$$

(Eq. 3-11)

Unit I DLH = First Unit DLH X U. F. (I)

where

Cum. F. = Cumulative Factor for the total number of units produced for the appropriate learning curve.

U. F. (I) = Unit Factor for Unit I for the appropriate learning curve.

These unit DLH can be generated by hand or by use of a computer and then used to calculate values for  $A_0$ ,  $A_1$ , and  $A_2$ . One such computer technique is described in reference 15. Once the unit DLH for each unit is determined, the DLH for the last unit produced will be used for  $X_1$  and the learning curve for the item produced will be used for  $X_2$  (.76, .80, .83, etc.).

When values are found for all the unknowns in the above equations, an approximation of the DLH for the first unit of a follow-on lot can be made. The total DLH for that lot can then be estimated by using the learning curve tables.

This method treats each production lot as if it were the first lot produced. Past learning is not taken into account when approximating a follow-on lot's first unit DLH. Thus, the above prediction technique provides good approximations for items where all learning is assumed lost due to a break in production. Also, if the last unit DLH is less than

50 hours then this model yields good results when using the regression constants calculated by Pichon and Richardson (see Eq. 4-3).

#### Model 2:

Another government agency, the Defense Contract Audit Agency (DCAA), also uses the learning curve as an estimating tool. Although DCAA is primarily an auditing agency, the organization also furnishes support to other government agencies in evaluating and negotiating government contracts. In performing this support activity, DCAA realized that a need existed for objectively measuring the learning lost due to a break in production. In efforts to satisfy this need, a project was initiated in 1971 to determine if such a task could be accomplished. As a result of this project, a study, directed by Mr. Robert B. Ilderton, was performed. As a result of this study, a method was developed whereby a weighted least-squares line is fitted, under the unit curve theory, to direct labor data before and after a production break in efforts to determine how many units are lost due to a break in production (11).

In developing this model, Mr. Ilderton also used simple calculus, linear regression analysis and the computer to approximate lot midpoints from cumulative data in order that an analysis of regression could be performed. Although the same tools were used in developing this model as were used to develop Model One, the end results were different.

Model Two is a modification of the basic learning curve and takes the form  $Y = K(X - AZ)^C$  where A equals the number of units of learning lost because of the break, Z equals zero before the break and 1 afterwards and the other variables are the same as the learning curve given in Chapter 2.

In using Mr. Ilderton's method an initial least-squares fit to the equation Y = KX<sup>C</sup> is accomplished in essentially the same way, whether labor hours are available for individual units or lots. Depending upon the amount of accuracy placed on the estimate, the data can be fitted in various ways. One way is to visually fit a curve through data points positioned on logarithmic graph paper using a straight edge to approximate a least-squares fit to these points. If a more accurate fit is desired, the data can be fitted to the equation:

where  $Y_X$  represents the logarithm of the average hours required to make units 1 through X.

Another alternative is that of fitting a curve to only two points:

(i) cumulative average hours through the last completed unit and (ii)

cumulative average hours through the completion of half that number of
units. The slope is equal to the first average divided by the second.

For better accuracy and faster estimating, a computer can be used to fit curves to historical data by applying simple linear regression formulae to the logarithms of the average hours and the number of units.

Once the learning curve has been determined, the parameter A in this model is set equal to 1 and a least-squares fit to the data is obtained after deducting one unit from all the units numbered after the break. The values obtained for the index of determination,  $(r^2)$ , from the two calculations are compared where

$$r^{2} = \frac{\left[N \sum (\log X \cdot \log Y) - \sum \log X \cdot \sum \log Y\right]^{2}}{\left[N \sum (\log X)^{2} - (\sum \log X)^{2}\right] \cdot \left[N \sum (\log Y)^{2} - (\sum \log Y)^{2}\right]}$$

$$= \frac{\text{Regression Sum of Squares}}{\text{Total Sum of Squares}}$$
(Eq. 3-12)

where

N = Number of units produced

X = Unit number

Y = DLH required to produce unit X

If the first index is greater, then no better fit is obtained from the model and no further calculations are required. If the second index is greater, the process is continued to provide fits to repositioned data with  $A = 2, 3, \ldots, 29, 30, 32, \ldots, 98, 100, 105, \ldots, 195, 200, 210, \ldots, 490, 500, 525, \ldots, 975, 1000, 1050, \ldots, 1950, 2000, 2100, \ldots, 4900, 5000, 5250, \ldots, 9750, 10,000, 10,500, \ldots until the values obtained for <math>r^2$  stop increasing and start decreasing. Thus, using the

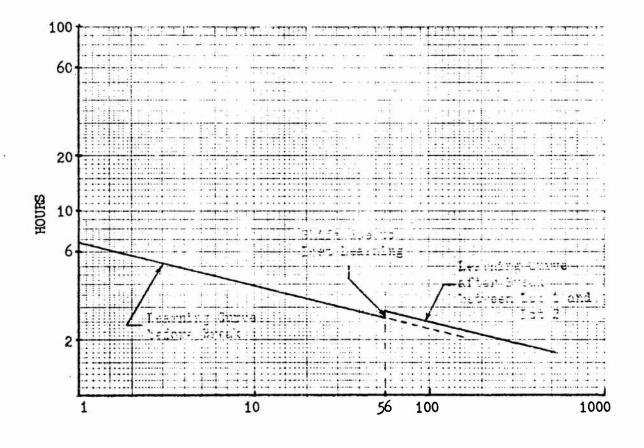
results from the least-squares fit which produced the highest value of  $r^2$ , the hours for the first unit following a break in production can be approximated. (11)

The above method treats each production lot as if it were a continuation of a past build. For this reason, past learning is used as a factor in predicting the DLH for the first unit of a follow-on production. When this method is applied, it is in essence shifting the learning curve up. Such a shift causes the unit cost of all units in a follow-on production to be larger when compared to the unit DLH without a break, as seen in Figure 3-2.

In viewing the results in a different light, the DLH for the first unit after a break would be equivalent to the cost of some previous unit. Thus, the break moves the starting position for the follow-on lot backward along the curve. Should all units be assumed lost, this method would yield approximately the same results as Method 1.

This method provides good approximations for the first unit of follow-on lots when learning is assumed to be retained after a break in production occurs.

The above methods for determining the direct labor hours after a break in production may be used for several purposes, some of which are listed below: (11)



UNITS

Figure 3-2. 85 Percent Learning Curve Before and After a Break in Production.

- a. A contractor has submitted a proposal for a follow-on procurement of an item for which there has been a break in production in the past. These methods can be used to determine the units of learning lost as a result of the break, reposition the units after the break and project the direct labor required for the proposed contract.
- b. A contractor has submitted a proposal for the follow-on procurement of an item which is not currently in production or for which a break in production will occur prior to the start of production under the proposed contract. It may be possible to estimate the loss of learning which will occur because of the break by using the above methods to determine the learning lost which occurred as a result of similar breaks in the production of similar items in the past.
- c. A contractor claims damages due to an interruption in production caused by Government error. These methods can be applied to normal data before and after the interruption to estimate the labor costs which the contractor would have incurred had there been no interruption.

It should be pointed out that computer programs have been developed for performing the calculations for both Method One and Method Two.

These programs can be found in reference 15 and 11 respectively.

#### CHAPTER 4

### APPLICATION OF MODELS

In efforts to show how the two methods presented in Chapter 3 can be applied, the following situation will be considered.

Suppose that a contract is to be let for 25 units of item X which has been previously built by the ABC Company. The first lot consisted of 25 units of item X which required 503.5 direct labor hours (DLH) to produce. Should the contract be awarded to the ABC Company, a break of 6 months will be realized by the time production could be resumed. With this in mind, the ABC Company submitted a proposal for 530 hours to build the follow-on 25 units. A local evaluation engineer has been asked to evaluate the contractor's proposal to determine if it is reasonable.

Since a six month production break will be realized, the evaluator should use an evaluation technique which considers the effects of a break in production upon future builds. Two such techniques will now be applied to the above situation.

### Method 1:

This approximation method applies when all learning is assumed to be lost due to a break in production. That is, no learning is retained from previous productions. Since the break is only six months, and the contractor still employs 98% of the employees that worked on the first

is retained. Thus, this method will not be applicable to our particular situation. However, the method will be illustrated to give the reader a procedure that can be followed when this method can be applied.

Recalling the model for the first method:

$$Y = e^{A_0 + A_1 X_1 + A_2 (\ln X_2)}$$
 (Eq. 4-1)

or

$$\ln Y = A_0 + A_1 X_1 + A_2 (\ln X_2)$$
 (Eq. 4-2)

where

Y = First unit DLH after a break in production  $X_1 = Last$  unit DLH for the lot preceding the break in production  $X_2 = Learning$  Curve (.83)

In order to apply this model, the following procedure is suggested.

Step 1.

The slope of the learning curve which applies to the item to be produced,  $X_2$ , must be determined. Since the prediction is for a follow-on production, actual data is available and appears in Appendix A. From this data, the learning curve which describes the production of item X is approximately an 83% curve, as shown in Appendix A. If, however, such data were not available, then the learning curve for similar items which the contractor has built should be used. If

there are no similar items, then the standard curve for the industry should be used for estimation purposes.

# Step 2.

Determine values for the constants  $A_0$  (the regression constant),  $A_1$ , and  $A_2$  (regression coefficients). Values for these constants have been calculated by Pichon and Richardson and their values are: (15)

$$A_0 = 1.09948$$
 $A_1 = 0.06020$ 
 $A_2 = -7.95450$ 

Since the last unit of item X produced requires less than 50 DLH, these constants will be used in this illustration (see Chapter 3). Thus, Eq. 4-2 becomes

$$\ln Y = 1.09948 + 0.06020 X_1 - 7.95450 (ln X_2)$$
 (Eq. 4-3)

Had the unit DLH been greater than 50 hours, then values for the above constants would be calculated using the method described in Chapter 3.

## Step 3.

The last piece of information needed to apply the model above is the direct labor hours for the last unit of the preceding lot,  $X_1$ . Since actual unit DLH are available, the DLH recorded for the last unit produced, 15.5 hours, will be used for  $X_1$ . If, however, the

unit DLH were not available, then the learning curve tables could be used to estimate the DLH for the last unit produced. Since the learning curve has been determined and the total DLH are known, the DLH for the last unit produced could be estimated as follows:

Last Unit DLH = 
$$\frac{\text{Total DLH}}{\text{Cum. F.}}$$
 X U. F.

where

Cum. F. = Cumulative Factor for the total number of units produced for the 83% learning curve

U. F. = Unit Factor for the last unit produced for the 83% learning curve

## Step 4.

Determine the first unit DLH for Lot 2.

Thus,

$$Y = 33.607$$
 Hours for Unit 1 of Lot 2

#### Step 5.

Now that the DLH for the first unit of the follow-on lot is known, the total DLH for the follow-on lot can be estimated. The estimated total DLH for the 25 units to be produced is

\* Cumulative Factor for 25 units using an 83% learning curve.

Since there is no set rule for determining reasonableness, the discretion lies with the evaluator. Thus, the contractor's proposal would be considered reasonable by this evaluator had all learning been lost due to the break in production. The reasonableness would be justified since the proposal and estimate differ only by 67.7 total hours, or 2.71 hours per unit.

#### Method 2:

This approximation method applies when learning is assumed to be retained from previous productions even though a break in production occurs. Since the ABC Company has retained 98% of the employees that worked on the first build of item X, learning is assumed to be retained. Thus, this method will be used to determine the reasonableness of the contractor's proposal.

Recalling the model for this method:

$$Y = K(X - AZ)^{C}$$
 (Eq. 4-4)

where

X = Number of hours to produce the first unit

X = Unit number to be produced

A = Number of units of learning lost because of the break

Z = 0 before the break and 1 after the break

c = log B/log 2 = log .83/log 2 = -.26882

In order to apply this model, the following procedure is suggested.

## Step 1.

The slope of the learning curve which applies to item X is determined as in Step 1 of Method 1. Thus, the 83% learning curve will be used.

#### Step 2.

Once the slope of the learning curve is determined, a value for c can be calculated by Eq. 2-3 from Chapter 2.

$$c = \frac{\text{Log S}}{\text{Log 2}}$$

Where S is the slope of the learning curve, .83. Thus, c takes on a value of -0.26882 for this evaluation.

# Step 3.

A value for K, the DLH for the first unit produced, can be calculated by using Eq. 3-11.

First Unit DLH = 
$$\frac{\text{Total DLH}}{\text{Cum. F.}}$$

In situations where actual data is available, as in our case, the actual DLH can be used. However, when the actual DLH for the first unit built differs from the value of K as calculated by Eq. A-2 in Appendix A, then the calculated value should be used. The calculated value would yield an estimate which conforms to the learning curve and reflects a more realistic value since it is calculated from DLH for all units produced. Also, the actual total hours for the first

unit could include time spent correcting unforeseen problems that developed as well as production time. For these reasons, the value for K to be used is 36.328 (see Appendix A).

# Step 4.

The next unknown, Z, will be equal to one in our evaluation since a break in production is expected.

# Step 5.

The unknown yet to be determined is A, the number of units of learning lost due to a break in production. In efforts to determine the value for A to be used, r<sup>2</sup> was calculated, using Eq. 3-12, from the least-squares fit to the equation  $Y = KX^{C}$ . In calculating the index of determination,  $r^2$ , N = 25, A = 0, Z = 1, X takes on values from 26 through 50, and Y equals the DLH for unit X. The resulting value obtained for the index was .9998 (see Appendix A). Then a least-squares fit was obtained for  $Y = K(X - AZ)^{c}$  setting A = 1, Z = 1, N = 25, X and Y are the same as above. The resulting  $r^2$  for this fit was .9918 which is less than the previous index (see Appendix A). Therefore, the value for A to be used in approximating the first unit of Lot 2 is zero. Thus, unit one of Lot 2, unit 26 to be produced, should require approximately 16.15 hours of production time as shown below.

$$Y = K(X - AZ)^{c}$$
  
= 36.328 (26 - 0 · 1)<sup>-0.26882</sup>

= 16.150 Hours to produce the 26th unit of item X

Had the first index been smaller than the second, then the procedure would be continued to provide least-squares fits to the repositioned data incrementing A by one from 2 through 30, by two from 30 through 100, by five from 100 through 200, etc., until such time as the value obtained for  $r^2$  stops increasing and begins to decrease. The value for A then, is that value which produces the highest value for  $r^2$ . Since the task of determining the largest value for  $r^2$  could be quite time-consuming, a computer should be used to perform the calculations. A copy of a program developed by Robert B.Ilderton of the Defense Contract Audit Agency which handles situations where breaks occur is included in Appendix B.

## Step 6.

Once the DLH for the first unit of the follow-on lot are calculated, the total cost for that lot can be approximated. The total DLH for the follow-on lot will be the sum of the hours for the units from X - A + 1 to X - A + N inclusive, where X is the last unit produced, A is the number of units lost and N is the number of units to be produced. The total DLH can be approximated using the following equation.

(Cum. F. L. U. - Cum. F. F. U.) X N = Total DLH (Eq. 4-5) where

Cum. F. L. U. = Cumulative Factor for last unit (Unit X - A + N)
Cum. F. F. U. = Cumulative Factor for first unit (Unit X - A + 1)

N = Number of units in follow-on lot

Using the learning curve tables for an 83% curve and Eq. 4-5, the total DLM should be approximately 226 hours as shown here.

 $(23.2198 - 14.1727) \times 25 = 226.178 \text{ Hours}$ 

In comparing the contractor's proposed DLH to the estimated DLH above, the 530 hours proposed by the ABC Company are 2.34 times the estimated hours. Thus, the contractor's proposed hours are too high and it is recommended that the contractor's proposal be rejected.

# Comparison:

In comparing the two methods presented in this chapter, there are some things that should be considered before applying either of the two models. For Method 1 to be applicable, there should exist a situation in which it can be assumed that no learning is retained whenever a break in production occurs. Also, one model does not work for all production items. For instance, when using the model developed by Pichon and Richardson, the accuracy deteriorates as the DLH of the items being evaluated increase. Their model appears to yield reasonable estimates

for items requiring up to approximately 50 hours of production but begins to worsen as the costs increase beyond that point. Thus, for higher cost items, a new equation would have to be generated.

For the second method, it is assumed that learning is passed on from lot to lot. That is, learning obtained from previous productions is retained for follow-on productions. Thus, when learning is assumed to be passed from one lot to another, this prediction method will yield good approximations for direct labor hours. If, however, all learning is assumed to be lost for one reason or another, then this method will produce approximately the same results as Method 1.

#### CHAPTER 5

## CONCLUSIONS AND RECOMMENDATIONS

In concluding this report, it is the opinion of this author that both prediction methods described and illustrated earlier in this report provide relatively close approximations for first units following a break in production, provided the proper assumptions are made. These again are: Method 1 presented an equation in which it was assumed that all learning is lost when a break is encountered. Thus, the estimated direct labor hours for the first unit following a break in production is made as though the follow-on production is the first. Put another way, the first unit of the follow-on production is treated as though it were the first unit to be produced. Method 2 presented an equation which predicts DLH assuming that learning can be retained from previous production runs. Even when a break in production occurs, some, but generally not all, learning may be retained.

The two models analyzed in this research paper are the only two models found by this author that attempt to incorporate a break in production into the approximation of follow-on productions. Thus, since breaks so often occur in product production, there is a great need for research in the area of production breaks and the effects they have on follow-on productions. Therefore, additional approximation techniques are needed that can be used to estimate the amount of learning lost due to

production breaks. One approach might be to set up experiments whereby data is collected on various production products in which breaks of varying lengths are experienced. This data could be used to formulate other models which would possibly yield more realistic and more accurate results.

It is the hope of this author that additional research will be conducted in the area of breaks in production and new and better approximation techniques will be developed.

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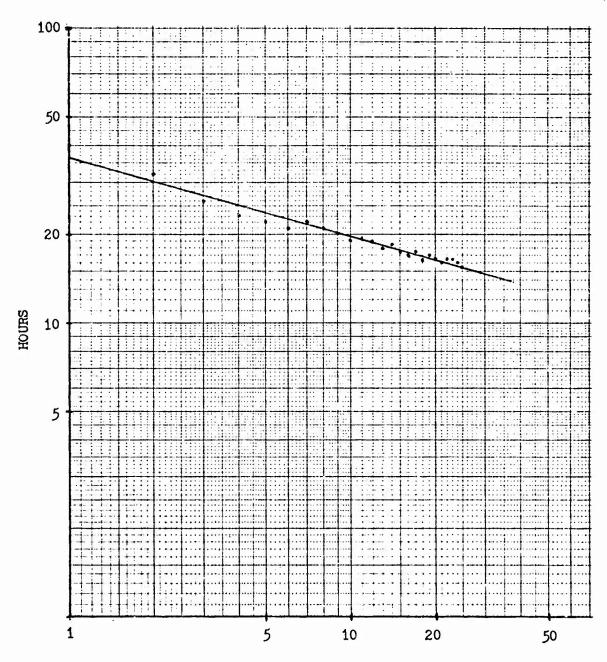
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# APPENDIX A

# DATA FOR APPLYING BOTH PREDICTION MODELS DESCRIBED IN CHAPTER 3

# Lot #1:

Item No.	Direct Labor Hours Y;
1	40.0
2 .	32.0
3	26.0
4	23.0
5	22.0
6	21.0
7	22.0
8	21.0
9	20.0
10	19.0
11	19.5
12	19.0
13	18.0
14	18.5
15	17.5
16	17.0
17	17.5
18	16.5
19	17.0
20	16.5
21	16.0
22	16.5
23	16.5
24	16.0
25	15.5



UNITS

Figure A-1. Line fitted to the actual data from Lot 1 using logarithmic graph paper.

$$N = 25$$

$$c^* = \frac{N \cdot \Sigma \left(\text{Log } X \cdot \text{Log } Y\right) - \Sigma \text{Log } X \cdot \Sigma \text{Log } Y}{N \cdot \Sigma \left(\text{Log } X\right)^2 - \left[\Sigma \text{Log } X\right]^2}$$
 (Eq. A-1)

$$c = -0.26623$$

$$\log K* = \frac{\sum \log Y - B \cdot \sum \log X}{N}$$
 (Eq. A-2)

Log K = 1.56025

Thus,

% Slope\* = 2° X 100% = 83.149% rounded to 83% so the same factor as used in the learning curve tables will be used.

Line fitted to the above data takes the form of

$$Log Y* = Log K + c \cdot Log X$$
 (Eq. A-3)

where

$$K = 36.328$$
;  $c = -0.26882$  (Factor for 83% Learning Curve) and  $X = X$ ;

The DLH for the first unit of Lot #1 is

$$\log Y = \log 36.328 + (-0.26882) \log 1$$

$$= 1.56025$$

Thus,

Y = 36.328 Hours for the first unit produced in Lot #1

\* See Reference (8)

The DLH for the last unit for Lot #1 is

$$Log Y = Log 36.328 + (-0.26882) Log 25$$

$$= 1.18908$$

Thus,

×.

Y = 15.45526 Hours for the last unit produced in Lot #1

# Calculation of Index of Determination (r<sup>2</sup>):

$$r^{2} = \frac{\left[N \sum \log X \cdot \log Y - \sum \log X \cdot \sum \log Y\right]^{2}}{\left[N \sum (\log X)^{2} - (\sum \log X)^{2}\right] \cdot \left[N \sum (\log Y)^{2} - (\sum \log Y)^{2}\right]}$$

where

$$N = 25$$
  
 $Y = 36.328 (X - A \cdot 1)^{-.26882}$   
 $X = Unit number (26 through 50)$ 

For A = 0:

$$\sum \log X \cdot \log Y = 44.656390$$

$$\sum \log X = 39.292430$$

$$\sum (\log X)^2 = 61.934727$$

$$\sum \log Y = 28.443447$$

$$\sum (\log Y)^2 = 32.374117$$

Thus

$$r^2 = .99989$$

For A = 1:

$$\sum \log X \cdot \log Y = 44.782180$$

$$\sum \log X = 39.292430$$

$$\sum (\log X)^2 = 61.934727$$

$$\sum \log Y = 28.524310$$

$$\sum (\log Y)^2 = 32.559201$$

Thus

$$r^2 = .99181$$

# APPENDIX B

This appendix containes a computer program developed by Robert B. Ilderton that fits weighted least-squares equations to the models  $Y = AX^B$  and  $Y = A(X-CZ)^B$ 

```
TO REM R. ILDERTON
 702DIMF(200), L(200), Y(200), M(200)
 704DATA9E9,9E9
 706READD, C
 7081FC=9E9THEN 1006
 710LETC=D+C
 7121FC>200THEN1002
 7141FC>=3THEN720
 716PRINT"THERE MUST BE AT LEAST THREE POINTS OF DATA."
 718ST0P
 720LETL=0
 722LETT=0
 724LETX 1=0
 726LETX 2= 0
 728LETY 1=0
 730LETY 2= 0
 732LETZ=0
 734FORI = 1TO C
 736REA@F(I),L(I),Y(I)
 7371FF(1)>LTHEN740
 738PRINT"FIRST UNIT IN LOT"; I; "IS"; F(I); ". THIS DOES NOT EXCEED"
 739PRINTL; "WHICH IS THE LAST UNIT IN THE PRIOR LOT.
                                                          CHECK YOUR DATA"
  40LETL=L(I)
 ,411FF-INT(F)>OTHEN1002
  1421F1.> INT(1.) THEN 1002
1431FL<F(1)THEN1002
 744LETN=L-F(1)+1
 745LETT=T+N
 746LETX=LOG((F(I)+L)/2)
 748LETX 1=X1+N*X
 750LEPX2=X2+N*X*X
 752LETY=LOG(Y(I))
 754LETY1=Y1+N*Y
 756LETY2=Y2+N*Y*Y
 758LETZ=Z+N*X*Y
 760NEXTI
 762LETY3=Y1/T
 764LETY4=Y2-Y1*Y3
 766READA
 7681FA<>9E9THEN1002
 770LETB=(2-X1*Y1/T)/(X2-X1*X1/T)
 772LETS=0
 774FORH=0T09
 776G0SUB930
 7781FABS(B1-B)<.00001THEN784
 780LETB=B1
 782NEXTH
 "841 FB1 < OTHEN 788
 - 186PRINT"DATA YIELDS SLOPE OF MORE THAN 100 PC"
 787STOP
 #88PRINT
 790PRINT "LEAST-SQUARES FIT TO Y=AX'B"
792PRINT
```

```
794:GOSUB1048
• 796LETE1=E
 797LETS 1=1
                                                                     53
  '98LETS2=0
 '991FL(C)>99THEN801
 600LETS1= . 01
 801FORS=S1T029*S1STEPS1
 802G05UB836
 804NEXT S
 806F0RS=30*S1T098*S1STEP2*S1
 808G0SUB836
 81 ONEXTS
 B12LETS1=S1*5
 814F0RS=20*S1T039*S1STEPS1
 816G0SUB836
 BIBNEXTS
 820F0RS=40*S1T098*S1STEP2*S1
 822G0SUB836
 824NEXTS
 826F0RS=100*S1T0195*S1STEP5*S1
 828G0 SUB836
 830NEXTS
 832LETS1=S1*10
 834G0T0814
.8361FS<F(D+1)-1THEN840
 838LETS=F(D+1)-1
 840G0SUB930
 8421FS=F(D+1)-1THEN862
  141FE/E1<.9999THEN856
MGIFE < EITHEN854
 46LETEI-E
 850LETB=B1
 852LETS2=S
 854RETURN
 856LETS=INT(100*S2+.5)/100
 8581FS=OTHEN874
 860G0SUB930
 862PRINT"LEAST-SQUARES FIT TO Y=A(X-CZ) + B, WHERE Z=O BEFORE"
 864PRINT"BREAK IN PRODUCTION AND 1 AFTER"
 866PRINT
 868 PRINT"C=", 5
 870G0 SUB1 048
 872G0T0878
 874PRINT'NO BETTER FIT IS OBTAINED FROM THE MODEL"
 876PRINT"Y=A(X-CZ) 1 B"
 878 LET C5=C
 880 LET C=1
 882LETI=1
 884PRINT
```

884PRINT
886PRINT"WHEN A QUESTION MARK APPEARS, TYPE THE FIRST (F) AND LAST (L)"
888PRINT"UNITS OF A LOT FOR WHICH CALCULATION OF PROJECTED HOURS OR"
890PRINT"COST IS DESIRED. WHEN NO MORE CALCULATIONS ARE NEEDED,"

'SPRINT"ENTER '0,0'."

```
54
895PRINT"F, L=";
8961NPUTF1, L(1)
8971FF1+L(1)=OTHEN1084
8981FF1-INT(F1)>OTHEN1002
8991FL(1)-INT(L(1))>OTHEN1002
9001FF1<1THEN1002
9011FL(1) < F1THEN1002
902LETF(1)=F1
9041FF1<LTHEN912
906F(1)=F(1)-S
908L(1)=L(1)-S
91260SUB930
914PRINT"MI DPOINT="," ",M(1)
916LETU=A*M(1)+B1
9181FF1<=L(C5)THEN924
920PRINT "REPOSITIONED VALUE OF F=",F(1)
922PRINT "REPOSITIONED VALUE OF L=",L(1):
924PRINT"PROJECTED UNIT VALUE=",U
926PRINT"PROJECTED TOTAL VALUE=",U*N
928G0T0894
930LETX 1=0
932LETX 2= 0
934LETZ=0
936LETW=0
938FORI=1TOC
940LETF=F(1)
942LETL=L(1)
9441F1<=DTHEN950
946LETF=F-S
948LETL=L-S
950LETN=L-F+1
952LETW=0
954FORK=FTOL
9561FK > 50THEN 962
958LETW=W+K+B
960NEXTK
961G0T0964
962LETW=W+((L+.5)*(1+B)-(K-.5)*(1+B))/(1+B)
964LETM(1)=(W/N)*(1/B)
9661FC=1THEN984
968LETX=LOG(M(I))
970LETX1=X1+N*X
972LETX2=X2+N*X*X
974LETZ=Z+N*X*LOG(Y(1))
976NEXTI
978LETZ1=Z-X1*Y3
980LETB1=Z1/(X2-X1*X1/T)
982LETE=B1*Z1
984RETURN
1002PRINT "DATA DOES NOT CONFORM TO PRESCRIBED FORMAT.
                                                         PLEASE CHECK
10041FC=1THEN894
1006PRINT"THIS PROGRAM FITS WEIGHTED LEAST-SQUARES EQUATIONS TO THE"
1008PRINT"MODELS Y=AX1B AND Y=A(X-CZ)1B, WHERE"
1010PRINT" Y=DIRECT LABOR HOURS OR COST PER UNIT"
1012PRINT" X=UNIT NUMBER OR LOT MIDPOINT"
```

1014PRINT" A=THEORETICAL UNIT 1 HOURS OR COST"
1016PRINT" B=IMPROVEMENT CURVE SLOPE COEFFICIENT"

```
C=POSITIVE INTEGER REIRESENTING UNITS OF LEARNING"
               LOST AS A RESULT OF A PRODUCTION BREAK"
1020PRINT"
1022PRINT"LINES 1 TO 699 ARE AVAILABLE FOR USE AS DATA STATEMENTS"
MO24PRINT"ENTER FIRST THE NUMBER OF LOTS PRIOR TO THE PRODUCTION BET
1026PRINT"THEN THE NUMBER AFTER THE BREAK, AND THEN THE FIRST UNIT,
1028 PRINT LAST UNIT AND AVERAGE HOURS OR COST PER UNIT FOR EACH LOT
1030PPINT"SEQUENCE.
                      TYPE 'RUN' ON THE NEXT LINE.
                                                       FOR EXAMPLE:"
:1032PRINT
              1 DATA 3,2"
1034PRINT"
:1036PRINT"
              11 DATA 1,1,1102"
              12 DATA 2,3,825"
1038PRINT"
              13 DATA 7,10,551.4"
1040PRINT"
1042PRINT"
              21 DATA 11,11,616"
              22 DATA 12, 16, 517"
:1044PRINT"
1046STOP
1048LETA=EXP(Y3-B1*X1/T)
1050PRINT "A=" A
1052PRINT "B=",B1
1054PRINT"PCT -=",100*2'B1
1056PRINT "INDEX=",E/Y4
1058 PRINT
1060PRINT"MIDPOINT", "CALCULATED Y", "ACTUAL Y", "PCT. DIFF."
1062PRINT
1064FORI=1TOC
1066LETK=A*M(1) + B1
1068PRINTM(1), K, Y(1),
1070LETP=INT(1000*(Y(I)/K-1)+.5)/10
10721FP<0PHEN1076
1074PRINT"
1076PRINTP
1078NEXT1
108 OPRINT
1082RETURN
1084PRINT
90000G0T090110
90010DISABLE ALL
90020AS=UID
90030B5="*"
'90040B$(2,6)=A$
90050FILE APPEND#1=BS
90060PRINT# IUSING90070, DAT, UID, PID, TIM
90070: *BREAK
                 #######
90080FILES
9009 OENABLE
90100RETURN
90110PRINT 'END OF JOB
```

READY

\*BREAK 05:09 07/26/73

THIS PROGRAM FITS WEIGHTED LEAST-SQUARES EQUATIONS TO THE MODELS Y=AX+B AND Y=A(X-CZ)+B, WHERE
Y=DIRECT LABOR HOURS OR COST PER UNIT
X=UNIT NUMBER OR LOT MIDPOINT
A=THEORETICAL UNIT 1 HOURS OR COST
B=IMPROVEMENT CURVE SLOPE COEFFICIENT
C=POSITIVE INTEGER REPRESENTING UNITS OF LEARNING
LOST AS A RESULT OF A PRODUCTION BREAK
ENTER FIRST THE NUMBER OF LOTS PRIOR TO THE PRODUCTION BREAK,
THEN THE NUMBER AFTER THE BREAK, AND THEN THE FIRST UNIT,
LAST UNIT AND AVERAGE HOURS OR COST PER UNIT FOR EACH LOT IN
SEQUENCE. TYPE 'SUN' ON THE NEXT LINE. FOR EXAMPLE:

1 DATA 3,2
11 DATA 1,1,1102
12 DATA 2,3,825
13 DATA 7,10,5514
21 DATA 11,11,616
22 DATA 12,16,517
NOW AT 1770
SRU'S:0.3

READY

```
TAPE
READ PAPER TAPE
TP ON
1 DATA5,11
2 DATA1,48867,.73831
3 DATA48868, 150566, . 42494
4 DATA150567,215367,.53933
5 DATA215368,355406,.38395
6 DATA355407,545124,.31550
10 DATA940276, 1086362, . 29361
11 DATA1086363,1324810,.23832
12 DATA: 324811, 1568678, . 22629
13 DATA1568679, 1816422, . 19670
14 DATA1816423,2135068,.18107
15 DATA2135059,2412150,.16989
16 DATA2412151,2763654,.15854
17 DATA2763655,3047023,.13867
18 DATA3047024, 3336169, . 14188
19 DATA3336170,3670508,.13856
20 DATA3670509,3954108,.13061
```

END OF PAPER TAPE INPUT READY

\*BREAK 05:16 07/26/73

LEAST-SQUARES FIT TO Y=AX+B

A= 33.0905 B= -.360149 PCT.= 77.9084 INDEX= .924082

MIDPOINT	CALCULATED Y	ACTUAL Y	PCT. DIFF.
14169.5	1.05827	.73831	-30.2
93412.1	•536556	. 42494	-20.8
181656.	.422266	•53933	27.7
281434.	.360672	. 38395	6.5
445685.	.305639	.3155	3.2
1 - 01213E+6	.227468	• 29361	29.1
1 . 2029 1E+6	.213752	. 23832	11.5
1.44441E+6	.200121	. 22629	13.1
1 . 6905E+6	· 18909B	. 1967	4
1.97282E+6	. 178867	. 18107	1.2
2.27169E+6	. 170007	. 16989	1
2.5852E+6	. 162273	. 15854	-2.3
2.90378E+6	. 155622	. 13867	-10.9
19009E+6	• 150439	. 14188	-5.7
50153E+6	.145476	.13856	-4.8
3.81112E+6	• 141104	.13061	-7.4

LEAST-SQUARES FIT TO Y=A(X-CZ)+B, WHERE Z=O BEFORE BREAK IN PRODUCTION AND I AFTER

C=	550000
A= .	37.5533
B=	376124
PCT.=	77.0505
INDEX=	•956135

MIDPOINT	CALCULATED Y	ACTUAL Y	PCT. DIFF.
13968 • 7	1.03647	.73831	-28.8
93336.3	• 507332	.42494	-16.2
181641.	. 39494	•53933	36.6
281388 •	• 334991	• 38395	14.6
445631.	. 281794	.3155	12
460662.	<b>.</b> 2783	.29361	5.5
650573.	. 244415	.23832	-2.5
892924.	.216972	• 22629	4.3
1.13946E+6	. 19796	• 1967	6
1.42165E+6	.182152	.18107	6
1.72105E+6	• 169518'·	•16989	• 2
2.03442E+6	.159161	.15854	4
2.35338E+6	• 150696	.13867	<b>-8</b>
2.63978E+6	.144325	-14188	-1.7
2.95117E+6	.138397	·13856	• 1
3.26088E+6	• 133299	• 13061	- 2

WHEN A QUESTION MARK APPEARS, TYPE THE FIRST (F) AND LAST (L) UNITS OF A LOT FOR WHICH CALCULATION OF PROJECTED HOURS OR COST IS DESIRED. WHEN NO MORE CALCULATIONS ARE NEEDED, ENTER '0,0'.

# F,L=?545125,2080000 MIDPOINT=

PROJECTED UNIT VALUE=
PROJECTED TOTAL VALUE=

1.19926E+6 .194188 298055.

# F.L=75080001,7580000

MIDPOINT=
PROJECTED UNIT VALUE=
PROJECTED TOTAL VALUE=

5.71727E+6 .107921 269803.

F.L=?0.0